## PROBABILITY AND SAMPLING

## Probability

Probability provides the language of data analysis.

Equiprobable outcomes definition:
Probability of event $E$ is number of outcomes favorable to $E$ divided by the total number of outcomes. e.g. Probability of a head $=1 / 2$.

Long-run frequency definition:
If event $E$ occurs $n$ times in $N$ identical experiments, the probability of $E$ is the limit of $n / N$ as $N$ goes to infinity.

Subjective probability:
Probability is a measure of belief.

## First Law of Probability

Law says that probability can take values only in the range zero to one and that an event which is certain has probability one.

$$
\left\{\begin{array}{l}
0 \leq \operatorname{Pr}(E) \leq 1 \\
\operatorname{Pr}(E \mid E)=1 \text { for any } E
\end{array}\right.
$$

i.e. If event $E$ is true, then it has a probability of 1 . For example:

$$
\operatorname{Pr}(\text { Seed is Round|Seed is Round })=1
$$

## Second Law of Probability

If $G$ and $H$ are mutually exclusive events, then:

$$
\operatorname{Pr}(G \text { or } H)=\operatorname{Pr}(G)+\operatorname{Pr}(H)
$$

For example,

$$
\operatorname{Pr}(\text { Seed is Round or Wrinkled })=\operatorname{Pr}(\text { Round })+\operatorname{Pr}(\text { Wrinkled })
$$

More generally, if $E_{i}, i=1, \ldots r$, are mutually exclusive then

$$
\begin{aligned}
\operatorname{Pr}\left(E_{1} \text { or } \ldots \text { or } E_{r}\right) & =\operatorname{Pr}\left(E_{1}\right)+\ldots+\operatorname{Pr}\left(E_{r}\right) \\
& =\sum_{i} \operatorname{Pr}\left(E_{i}\right)
\end{aligned}
$$

## Complementary Probability

If $\operatorname{Pr}(E)$ is the probability that $E$ is true then $\operatorname{Pr}(\bar{E})$ denotes the probability that $E$ is false. Because these two events are mutually exclusive

$$
\operatorname{Pr}(E \text { or } \bar{E})=\operatorname{Pr}(E)+\operatorname{Pr}(\bar{E})
$$

and they are also exhaustive in that between them they cover all possibilities - one or other of them must be true. So,

$$
\begin{aligned}
\operatorname{Pr}(E)+\operatorname{Pr}(\bar{E}) & =1 \\
\operatorname{Pr}(\bar{E}) & =1-\operatorname{Pr}(E)
\end{aligned}
$$

The probability that $E$ is false is one minus the probability it is true.

## Third Law of Probability

For any two events, $G$ and $H$, the third law can be written:

$$
\operatorname{Pr}(G \text { and } H)=\operatorname{Pr}(G) \operatorname{Pr}(H \mid G)
$$

There is no reason why $G$ should precede $H$ and the law can also be written:

$$
\operatorname{Pr}(G \text { and } H)=\operatorname{Pr}(H) \operatorname{Pr}(G \mid H)
$$

For example
$\operatorname{Pr}($ Seed is round $\&$ is type $A A)$
$=\operatorname{Pr}($ Seed is round $\mid$ Seed is type $A A) \times \operatorname{Pr}($ Seed is type $A A)$

$$
=1 \times p_{A}^{2}
$$

## Independent Events

If the information that $H$ is true does nothing to change uncertainty about $G$, then

$$
\operatorname{Pr}(G \mid H)=\operatorname{Pr}(G)
$$

and

$$
\operatorname{Pr}(H \text { and } G)=\operatorname{Pr}(H) \operatorname{Pr}(G)
$$

Events $G, H$ are independent.

## Law of Total Probability

If $G, \bar{G}$ are two mutually exclusive and exhaustive events $(\bar{G}=$ not $G$ ), then for any other event $E$, the law of total probability states that

$$
\operatorname{Pr}(E)=\operatorname{Pr}(E \mid G) \operatorname{Pr}(G)+\operatorname{Pr}(E \mid \bar{G}) \operatorname{Pr}(\bar{G})
$$

This generalizes to any set of mutually exclusive and exhaustive events $\left\{S_{i}\right\}$ :

$$
\operatorname{Pr}(E)=\sum_{i} \operatorname{Pr}\left(E \mid S_{i}\right) \operatorname{Pr}\left(S_{i}\right)
$$

For example

$$
\begin{aligned}
\operatorname{Pr}(\text { Seed is round })= & \operatorname{Pr}(\text { Round } \mid \text { Type } A A) \operatorname{Pr}(\text { Type AA }) \\
& +\operatorname{Pr}(\text { Round } \mid \text { Type Aa) } \operatorname{Pr}(\text { Type Aa }) \\
& +\operatorname{Pr}(\text { Round } \mid \text { Type aa }) \operatorname{Pr}(\text { Type aa }) \\
= & 1 \times p_{A}^{2}+1 \times 2 p_{A} p_{a}+0 \times p_{a}^{2} \\
= & p_{A}\left(2-p_{A}\right)
\end{aligned}
$$

## Bayes' Theorem

Bayes' theorem relates $\operatorname{Pr}(G \mid H)$ to $\operatorname{Pr}(H \mid G)$ :

$$
\begin{aligned}
\operatorname{Pr}(G \mid H) & =\frac{\operatorname{Pr}(G H)}{\operatorname{Pr}(H)}, \text { from third law } \\
& =\frac{\operatorname{Pr}(H \mid G) \operatorname{Pr}(G)}{\operatorname{Pr}(H)}, \text { from third law }
\end{aligned}
$$

If $\left\{G_{i}\right\}$ are exhaustive and mutually exclusive, Bayes' theorem can be written as

$$
\operatorname{Pr}\left(G_{i} \mid H\right)=\frac{\operatorname{Pr}\left(H \mid G_{i}\right) \operatorname{Pr}\left(G_{i}\right)}{\sum_{i} \operatorname{Pr}\left(H \mid G_{i}\right) \operatorname{Pr}\left(G_{i}\right)}
$$

## Bayes' Theorem Example

Suppose $G$ is event that a man has genotype $A_{1} A_{2}$ and $H$ is the event that he transmits allele $A_{1}$ to his child. $\operatorname{Then} \operatorname{Pr}(H \mid G)=$ 0.5 .

Now what is the probability that a man has genotype $A_{1} A_{2}$ given that he transmits allele $A_{1}$ to his child?

$$
\begin{aligned}
\operatorname{Pr}(G \mid H) & =\frac{\operatorname{Pr}(H \mid G) \operatorname{Pr}(G)}{\operatorname{Pr}(H)} \\
& =\frac{0.5 \times 2 p_{1} p_{2}}{p_{1}} \\
& =p_{2}
\end{aligned}
$$

## Sampling

Statistical sampling: The variation among repeated samples from the same population ("fixed" sampling). Inferences can be made about that particular population.

Genetic sampling: The variation among replicate (conceptual) populations ("random" sampling). Inferences are made to all populations with the same history.

## Classical Model

Reference population (Usually assumed infinite and in equilibrium)


## Coalescent Theory

An alternative framework works with genealogical history of a sample of alleles. There is a tree linking all alleles in a current sample to the "most recent common ancestral allele." Allelic variation due to mutations since that ancestral allele.

The coalescent approach requires mutation and may be more appropriate for long-term evolution and analyses involving more than one species. The classical approach allows mutation but does not require it: within one species variation among populations may be due primarily to drift.

